## Gluon: A Communication-Optimizing Substrate for Distributed Heterogeneous Graph Analytics

Roshan Dathathri Gurbinder Gill Loc Hoang Hoang-Vu Dang Alex Brooks Nikoli Dryden Marc Snir Keshav Pingali





### Distributed Graph Analytics

### **Applications:**

machine learning and network science



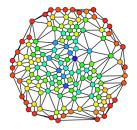






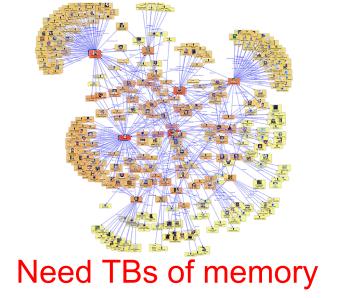






Credits: Wikipedia, SFL Scientific, MakeUseOf

### **Datasets:** unstructured graphs



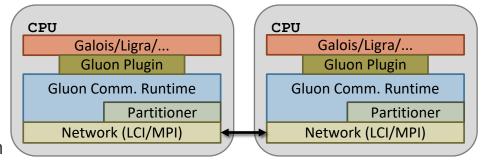
Credits: Sentinel Visualizer

### Gluon [PLDI'18]

- Substrate: single address space applications on distributed, heterogeneous clusters
- Provides:
  - Partitioner
  - High-level synchronization API
  - Communication-optimizing runtime

### How to use Gluon?

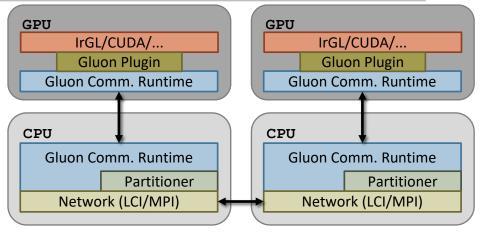
- Programmers:
  - Write shared-memory applications
  - Interface with Gluon using API
- Gluon transparently handles:
  - Graph partitioning
  - Communication and synchronization



Galois [SoSP'13] Ligra [PPoPP'13] IrGL [OOPSLA'16] LCI [IPDPS'18]

### How to use Gluon?

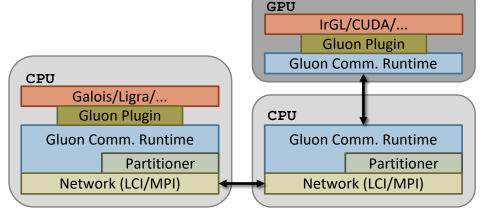
- Programmers:
  - Write shared-memory applications
  - Interface with Gluon using API
- Gluon transparently handles:
  - Graph partitioning
  - Communication and synchronization



Galois [SoSP'13] Ligra [PPoPP'13] IrGL [OOPSLA'16] LCI [IPDPS'18]

### How to use Gluon?

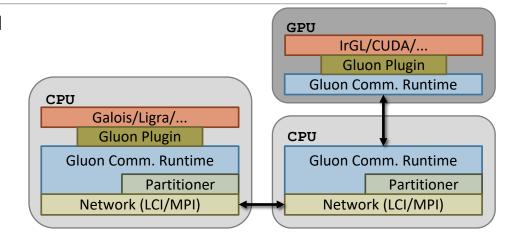
- Programmers:
  - Write shared-memory applications
  - Interface with Gluon using API
- Gluon transparently handles:
  - Graph partitioning
  - Communication and synchronization



Galois [SoSP'13] Ligra [PPoPP'13] IrGL [OOPSLA'16] LCI [IPDPS'18]

### Contributions

- Novel approach to build distributed and heterogeneous graph analytics systems out of plug-and-play components
- Novel optimizations that reduce communication volume and time
- Plug-and-play systems built with Gluon outperform the state-of-the-art



Galois [SoSP'13] Ligra [PPOPP'13] IrGL [OOPSLA'16] LCI [IPDPS'18]

### Outline

- Gluon Synchronization Approach
- Optimizing Communication
  - Exploiting Structural Invariants of Partitions
  - Exploiting Temporal Invariance of Partitions
- Experimental Results

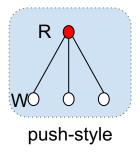
# Gluon Synchronization Approach

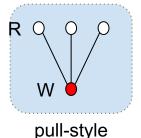
## Vertex Programming Model

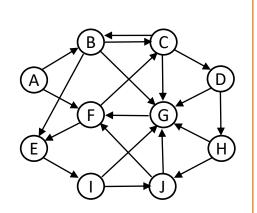
- Every node has one or more labels
  - e.g., distance in single source shortest path (SSSP)
- Apply an operator on an active node in the graph
  - e.g., relaxation operator in SSSP

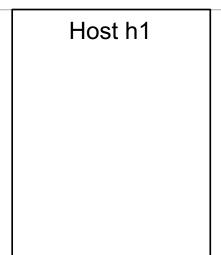


- Push-style: reads its label and writes to neighbors' labels
- Pull-style: reads neighbors' labels and writes to its label
- Applications: breadth first search, connected component, pagerank, single source shortest path, betweenness centrality, k-core, etc.





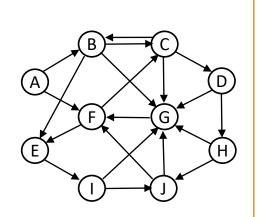


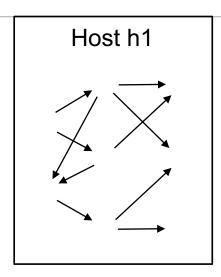


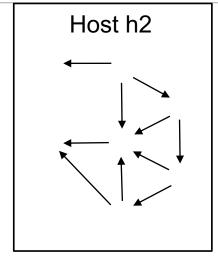
Host h2

Original graph

Partitions of the graph



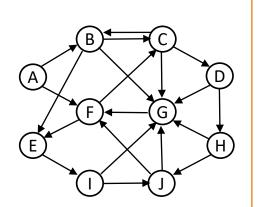


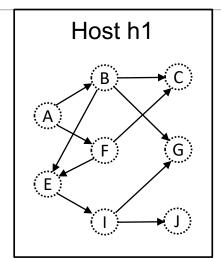


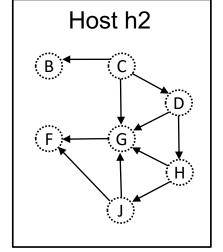
 Each edge is assigned to a unique host

Original graph

Partitions of the graph



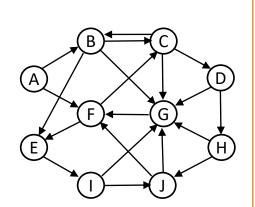




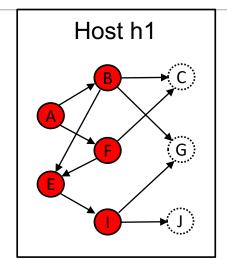
- Each edge is assigned to a unique host
- All edges
   connect proxy
   nodes on the
   same host

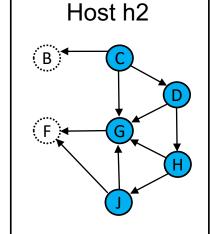
Original graph

Partitions of the graph



Original graph



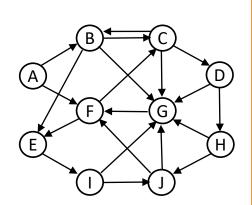


: Master proxy

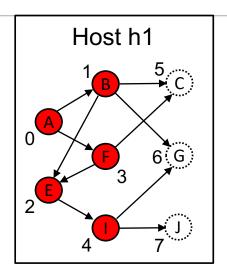
: Mirror proxy

Partitions of the graph

- Each edge is assigned to a unique host
- All edges
  connect proxy
  nodes on the
  same host
- A node can have multiple proxies: one is master proxy; rest are mirror proxies



Original graph



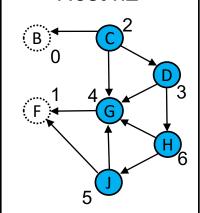
A-J: Global IDs

0-7: Local IDs

: Mirror proxy

Partitions of the graph

Host h2

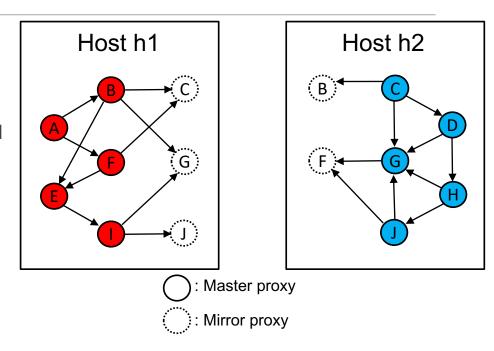


: Master proxy

- Each edge is assigned to a unique host
- All edges
  connect proxy
  nodes on the
  same host
- A node can have multiple proxies: one is master proxy; rest are mirror proxies

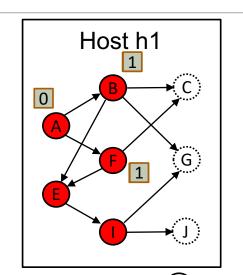
### How to synchronize the proxies?

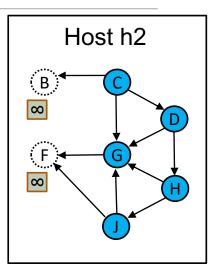
- Distributed Shared Memory (DSM) protocols
  - Proxies act like cached copies
  - Difficult to scale out to distributed and heterogeneous clusters



### How does Gluon synchronize the proxies?

- Exploit domain knowledge
  - Cached copies can be stale as long as they are eventually synchronized





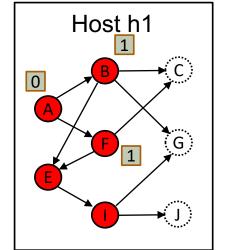
): Master proxy

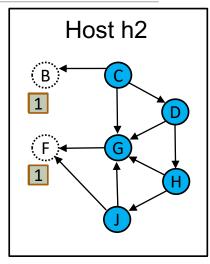
: Mirror proxy

: distance (label) from source A

## How does Gluon synchronize the proxies?

- Exploit domain knowledge
  - Cached copies can be stale as long as they are eventually synchronized
- Use all-reduce:
  - Reduce from mirror proxies to master proxy
  - Broadcast from master proxy to mirror proxies



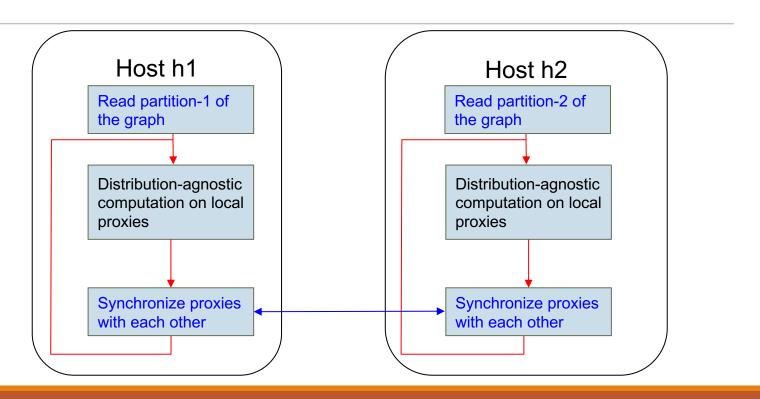


): Master proxy

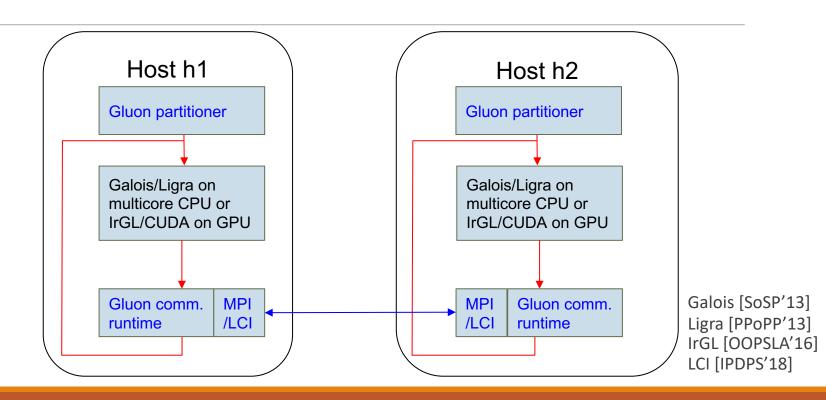
: Mirror proxy

: distance (label) from source A

### When to synchronize proxies?



### Gluon Distributed Execution Model



## Gluon Synchronization API

- Application-specific:
  - What: Label to synchronize
  - When: Point of synchronization
  - O How: Reduction operator to use
- Platform-specific:
  - Access functions for labels (specific to data layout)

# Exploiting Structural Invariants to Optimize Communication

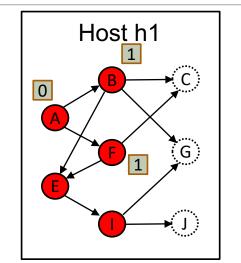
### Structural invariants in the partitioning

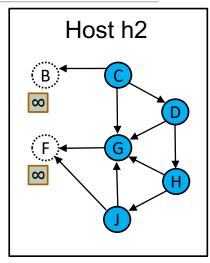
### Structural invariants in this partitioning:

 Mirror proxies do not have outgoing edges

### As a consequence, for sssp:

- Mirror proxies do not read their distance label
- Broadcast from master proxy to mirror proxies is not required



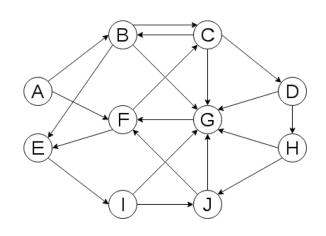


: Master proxy

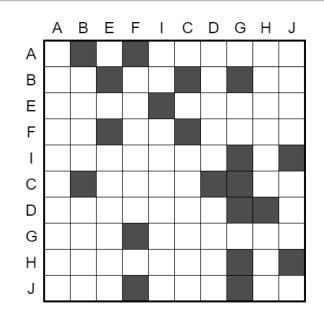
: Mirror proxy

: distance (label) from source A

## Graph as an adjacency matrix

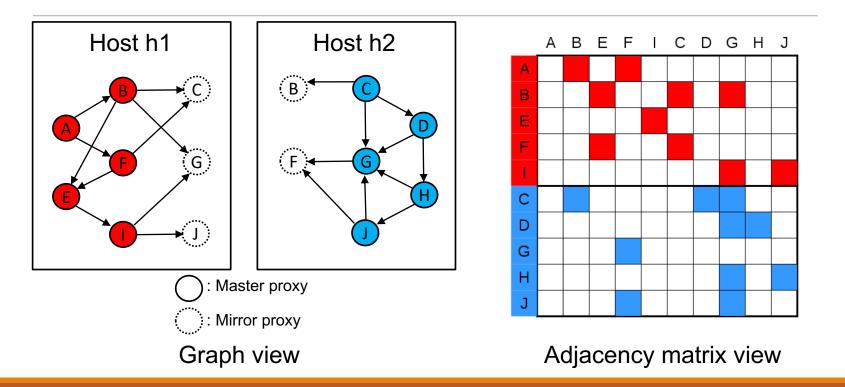


Graph view

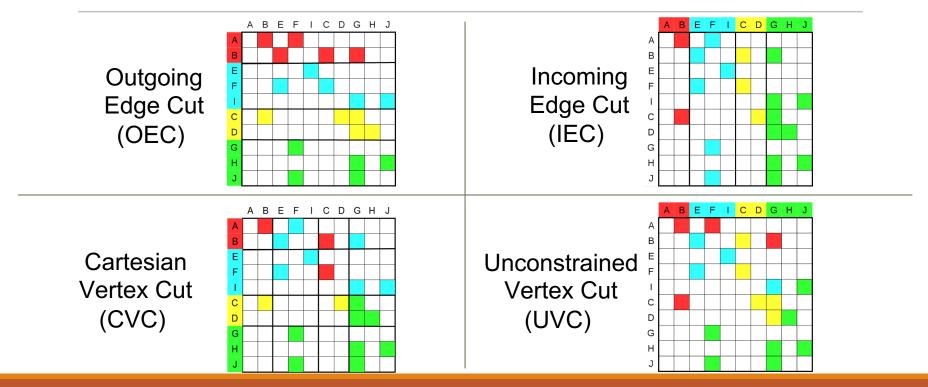


Adjacency matrix view

### Partitioning an adjacency matrix



### Partitioning strategies with 4 partitions



### Partitioning: strategies, constraints, invariants

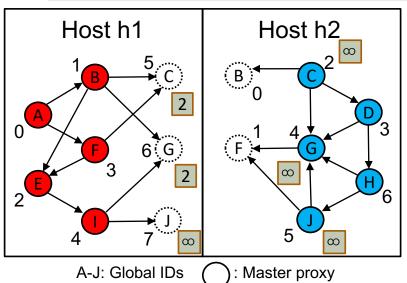
Algorithm invariant in SSSP:

R ● → W

Strategy	Constraints and Invariants	SSSP: Invariants	SSSP: Sync
Outgoing Edge-Cut (OEC)	Mirrors: no outgoing edges	Mirrors: label not read	Reduce
Incoming Edge-Cut (IEC)	Mirrors: no incoming edges	Mirrors: label not written	Broadcast
Cartesian Vertex-Cut (CVC)	Mirrors: either no outgoing edges or no incoming edges	Mirrors: either label not read or label not written	Reduce-partial & Broadcast-partial
Unconstrained Vertex-Cut (UVC)	None	None	Reduce & Broadcast

# Exploiting Temporal Invariance to Optimize Communication

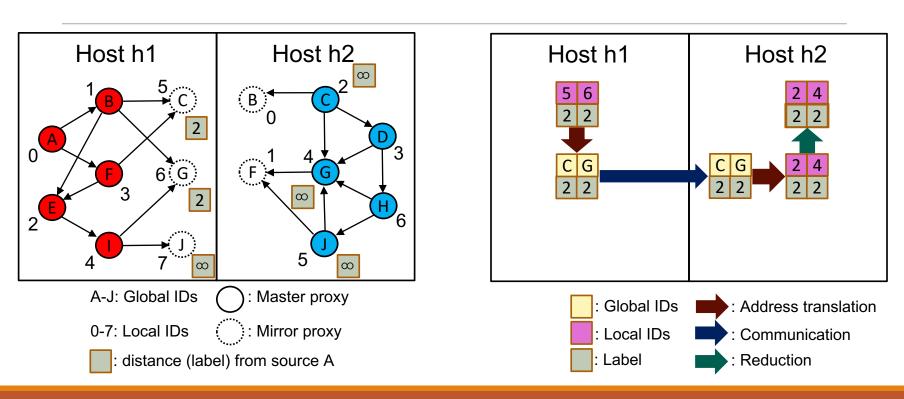
### **Bulk-communication**



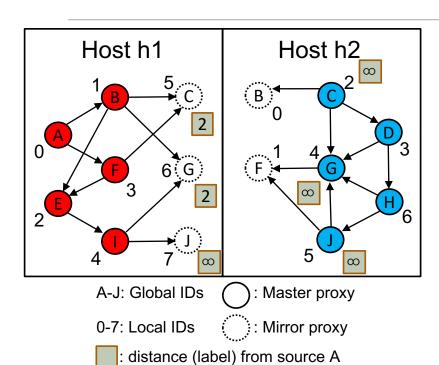
- 0-7: Local IDs : Mirror proxy
- distance (label) from source A

- Proxies of millions of nodes need to be synchronized in a round
  - Not every node is updated in every round
- Address spaces (local-IDs) of different hosts are different
- Existing systems: use address translation and communicate global-IDs along with updated values

### Bulk-communication in existing systems

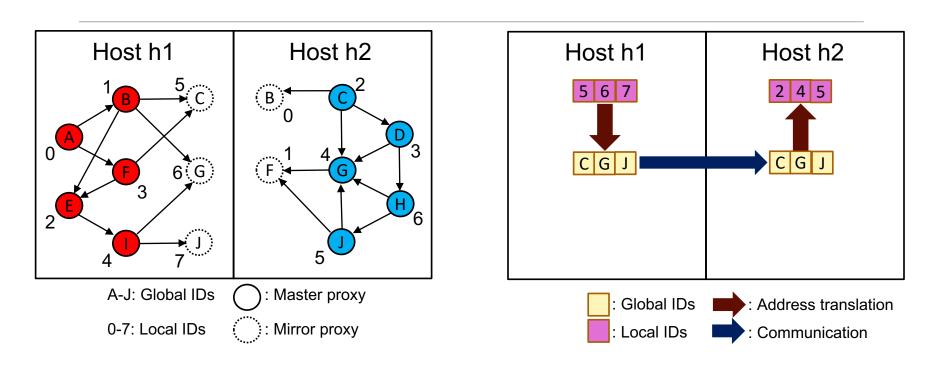


### Bulk-communication in Gluon



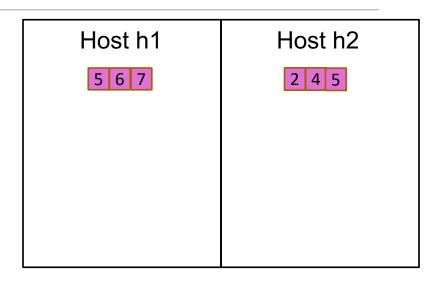
- Elides address translation during communication in each round
- Exploits temporal invariance in partitioning
  - Mirrors and masters are static
  - e.g., only labels of C, G, and J can be reduced from h1 to h2
- Memoize address translation after partitioning

### Optimization I: memoization of address translation



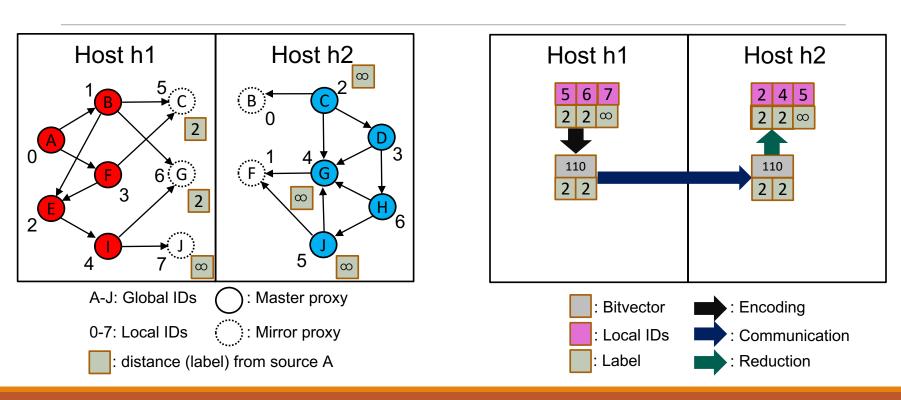
### Communication after memoization

- Memoization establishes the order of local-IDs shared between hosts
- Send and receive labels in each round in the same order
- Only some labels updated in a round:
  - Send simple encoding that captures which of the local-IDs in the order were updated



: Local IDs

### Optimization II: encoding of updated local-IDs



## Experimental Results

### Experimental setup

#### Systems:

- D-Ligra (Gluon + Ligra)
- D-Galois (Gluon + Galois)
- D-IrGL (Gluon + IrGL)
- Gemini (state-of-the-art) [OSDI'16]

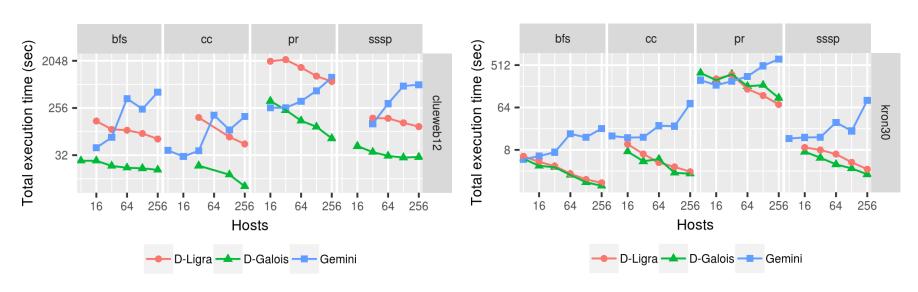
Inputs	rmat28	kron30	clueweb12	wdc12
V	268M	1073M	978M	3,563M
E	4B	11B	42B	129B
E / V	16	16	44	36
Size (CSR)	35GB	136GB	325GB	986GB

#### Benchmarks:

- Breadth first search (bfs)
- Connected components (cc)
- Pagerank (pr)
- Single source shortest path (sssp)

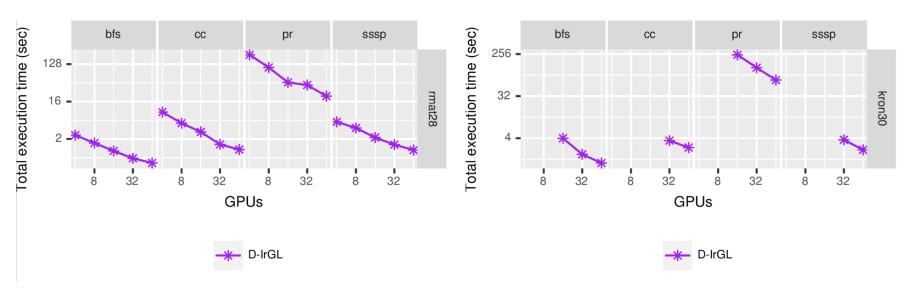
Clusters	Stampede (CPU)	Bridges (GPU)
Max. hosts	256	64
Machine	Intel Xeon Phi KNL	4 NVIDIA Tesla K80s
Each host	272 threads of KNL	1 Tesla K80
Memory	96GB DDR3	12GB DDR5

# Strong scaling on Stampede (68 cores on each host)



D-Galois performs best and scales well

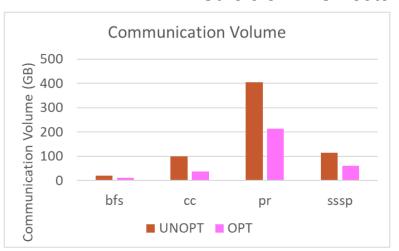
# Strong scaling on Bridges (4 GPUs share a physical node)

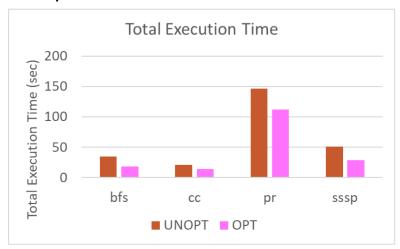


**D-IrGL** scales well

#### Impact of Gluon's communication optimizations

D-Galois on 128 hosts of Stampede: clueweb12



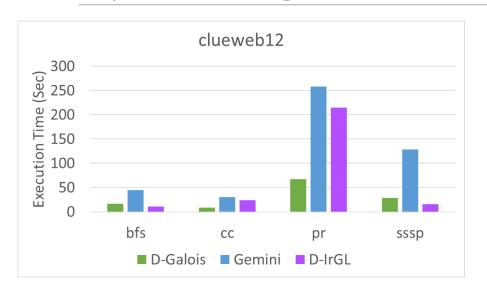


Improvement (geometric mean):

Communication volume: 2x

Execution time: 2.6x

### Fastest execution time (sec) using best-performing number of hosts/GPUs





D-Galois and D-IrGL are faster than Gemini by factors of 3.9x and 4.9x

#### Subsequent work using Gluon

- [EuroPar'18] **Abelian** compiler: shared-memory Galois apps ---> distributed, heterogeneous (D-Galois + D-IrGL) apps
- [VLDB'18] Partitioning Policies: Cartesian Vertex Cut performs best at scale
- [PPoPP'19] Min-Rounds Betweenness Centrality (MRBC) algorithm
- [ASPLOS'19] **Phoenix**: fault-tolerance without overhead during fault-free execution
- [IPDPS'19] Fast Customizable Streaming Edge Partitioner (CuSP)

#### Conclusions

- Novel approach to build distributed, heterogeneous graph analytics systems: scales out to 256 multicore-CPUs and 64 GPUs
- Novel communication optimizations: improve execution time by 2.6x
- Gluon, D-Galois, and D-IrGL: publicly available in Galois v4.0

http://iss.ices.utexas.edu/?p=projects/galois



Use Gluon to scale out your shared-memory graph analytical applications

### Backup slides

#### Graph construction time (sec)

1 host	rmat26	twitter40	rmat28	256 hosts	rmat28	kron30	clueweb12	wdc12
	271.6	158.3	396.9	D-Ligra			470.5	1515.9
Galois	64.9	51.8	123.9	<b>D-Galois</b>	65.5	225.7	396.2	1345.0
Gemini	854.3	893.5	3084.7	Gemini	231.0	921.8	1247.7	X

# Execution time (sec) on a single node of Stampede ("-" means out-of-memory)

Input		twi	tter40		rmat28			
Benchmark	bfs cc pr sssp		bfs	cc	pr	sssp		
Ligra	0.31	2.75	175.67	2.60	0.77	17.56	542.51	-
D-Ligra	0.44	3.16	188.70	2.92	1.21	18.30	597.30	-
Galois	0.68	2.73	43.47	5.55	2.54	13.20	116.50	21.42
<b>D-Galois</b>	1.03	1.04	86.53	1.84	4.05	7.02	326.88	5.47
Gemini	0.85	3.96	80.23	3.78	3.44	20.34	351.65	41.77

# Execution time (sec) on a single node of Bridges with 4 K80 GPUs ("-" means out-of-memory).

Input		rmat26				twitter40			
Benchmark	bfs	cc	pr	sssp	bfs	cc	pr	sssp	
Gunrock	-	1.81	51.46	1.42	0.88	1.46	37.37	2.26	
D-IrGL(OEC)	3.61	5.72	55.72	4.13	1.03	1.57	62.81	1.99	
D-IrGL(IEC)	0.72	7.88	7.65	0.84	0.73	1.55	35.03	1.44	
D-IrGL(HVC)	0.82	1.53	8.54	0.95	1.08	1.58	44.35	2.04	
D-IrGL(CVC)	2.11	4.22	46.91	2.24	0.87	1.39	46.86	2.32	

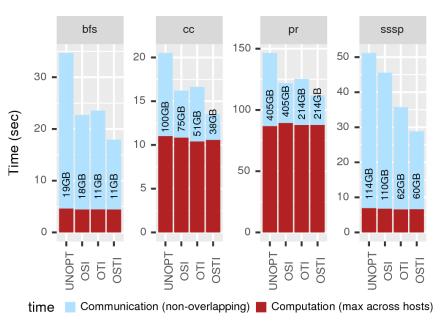
# Fastest execution time (sec) of all systems using best-performing number of hosts/GPUs

Bench-	Input		GPUs			
mark		D-Ligra D-Galois		Gemini   D-IrGL		
	rmat28	1.0 (128)	0.8 (256)	2.3 (8)	0.5 (64)	
bfs	kron30	1.6 (256)	1.4 (256)	5.0 (8)	1.2 (64)	
	clueweb12	65.3 (256)	16.7 (256)	44.4 (16)	10.8 (64)	
	wdc12	1995.3 (64)	380.8 (256)	X	-	
	rmat28	1.4 (256)	1.3 (256)	6.6 (8)	1.1 (64)	
cc	kron30	2.7 (256)	2.5 (256)	14.6 (16)	2.5 (64)	
	clueweb12	52.3 (256)	8.1 (256)	30.2 (16)	23.8 (64)	
	wdc12	176.6 (256)	75.3 (256)	X	-	

Bench- Input mark			CPUs				
		D-Ligra	D-Ligra D-Galois (		D-IrGL		
	rmat28	19.7 (256)	24.0 (256)	108.4 (8)	21.6 (64)		
pr	kron30	74.2 (256)	102.4 (256)	190.8 (16)	70.9 (64)		
	clueweb12	821.1 (256)	67.0 (256)	257.9 (32)	215.1 (64)		
	wdc12	663.1 (256)	158.2 (256)	X	-		
	rmat28	2.1 (256)	1.4 (256)	6.3 (4)	1.1 (64)		
sssp	kron30	3.1 (256)	2.4 (256)	13.9 (8)	2.3 (64)		
	clueweb12	112.5 (256)	28.8 (128)	128.3 (32)	15.8 (64)		
	wdc12	2985.9 (256)	574.9 (256)	X	-		

D-Galois and D-IrGL are faster than Gemini by factors of **3.9x** and **4.9x** on the average.

#### Impact of Gluon's communication optimizations



D-Galois on 128 hosts of Stampede: clueweb12 with CVC

Improvement (geometric mean):

Communication volume: 2x

• Execution time: **2.6x** 

#### Impact of Gluon's communication optimizations

#### D-Galois on 128 hosts of Stampede: clueweb12 with CVC

Bench	Commu volum		Total execution time (sec)		
mark	UNOPT	ОРТ	UNOPT	ОРТ	
bfs	19	11	34.7	17.9	
СС	100	38	20.5	13.5	
pr	405	214	146.5	111.8	
sssp	114	60	51.2	28.8	

#### Improvement (geometric mean):

Communication volume: 2x

• Execution time: **2.6x** 

## Fastest execution time (sec) of all systems using best-performing number of hosts/GPUs

clueweb12

Bench mark	D-Galois	Gemini	D-IrGL
bfs	16.7 (256)	44.4 (16)	10.8 (64)
СС	8.1 (256)	30.2 (16)	23.8 (64)
pr	67.0 (256)	257.9 (32)	215.1 (64)
sssp	28.8 (128)	128.3 (32)	15.8 (64)

D-Galois and D-IrGL are faster than Gemini by factors of 3.9x and 4.9x

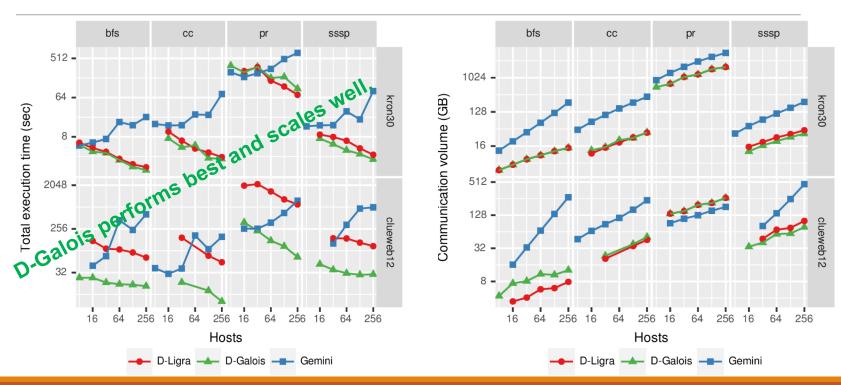
# Fastest execution time (sec) of all systems using best-performing number of hosts

wdc12

Bench mark	D-Ligra	D-Galois	Gemini
bfs	1995	381	Х
СС	177	75	Х
pr	663	158	Х
sssp	2986	575	Х

#### **D-Galois performs best**

# Strong scaling on Stampede (68 cores on each host)



#### Motivation

- Distributed CPU-only graph analytics:
  - Gemini [OSDI'16], PowerGraph [OSDI'12], ...
  - No way to reuse infrastructure, such as to leverage GPUs
- Decouple communication from computation
- Enable communication optimization at runtime